On the stress update algorithm of an advanced critical state elasto-plastic model and the effect of yield function equation

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A B S T R A C T

This paper investigates the computational performance of S-CLAY1S constitutive model by varying its yield function equation. S-CLAY1S is an advanced anisotropic elasto-plastic model that has been developed based on the extension of conventional critical state theory. In addition to modified Cam-Clay’s hardening law, S-CLAY1S also accounts for inherent and evolving plastic anisotropy, interparticle bonding and degradation of bonds during plastic straining. A modified Newton–Raphson stress update algorithm has been adopted for the implementation of the model and it was found that the algorithm’s convergence performance is sensitive to the expression of the yield function. It is shown that for an elasto-plastic model which is developed based on the critical state theory, it is possible to improve the performance of the numerical implementation by changing the form of the yield function. The results of this work can provide a new perspective for computationally cost-effective implementation of complex constitutive models in finite element analysis that can yield in more efficient boundary value level simulations.

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1. Introduction

It is a well-established fact in soil mechanics that the yield curves obtained from experimental tests (with triaxial or hollow cylinder apparatuses) on undisturbed samples of natural clays are inclined due to the inherent fabric anisotropy in the soil’s structure (e.g. [1–3]). Since consideration of full anisotropy in modelling soil’s behaviour is not practical due to the number of parameters involved, efforts have been mainly focused on the development of models with reduced number of parameters while maintaining the capacity of the model [4]. In order to capture the effects of anisotropy on soil’s behaviour, a number of researcher have proposed anisotropic elasto-plastic constitutive models involving an inclined yield curve that is either fixed (e.g. [5]) or is able to rotate in order to simulate the development or erasure of anisotropy during plastic straining (e.g. [6–9]). Among the developed anisotropic models, S-CLAY1S model [10] has been well-validated and accepted as a practical anisotropic model for simulating natural clays’ behaviour (e.g. [11–13]). The consideration of anisotropy in this model is similar to an earlier version of the model called S-CLAY1 [8]. In both S-CLAY1 and S-CLAY1S models the initial anisotropy is assumed to be cross-anisotropic, which is a realistic assumption for normally consolidated clays deposited along the direction of consolidation. Both models account for the development or erasure of anisotropy if the subsequent loading produces irrecoverable strains, resulting a generalised plastic anisotropy condition. S-CLAY1S model is further extended to account for inter-particle bonding in natural soil’s structure, and subsequent destructuration of bonds when the soil is under plastic straining. The main advantages of S-CLAY1S model over other proposed anisotropic models are (i) its realistic $K_0$ prediction, and most importantly (ii) the fact that model parameter values can be determined from standard laboratory tests using well-defined methodologies [10].

With additional complexity of a constitutive model, comes additional computational cost for its application. Particularly implementation of advanced constitutive models to solve boundary value problems requires integration algorithms that are not only stable, robust and accurate but also computationally efficient; because, in this case, the number of calculations involved per analysis increment is much higher than in the case of simple element level simulations for model validation. The main problem for successful implementation of complex constitutive models in the finite element method is error control that is the result of adopting finite number of loading steps. There are generally two techniques to solve a constitutive model’s system of equations at a stress point, the explicit (incremental) method and the implicit
Nomenclature

\[ a_0, \alpha \] initial and current value of anisotropy
\[ d, \Delta \] deviatoric fabric tensor
\[ d\alpha \] change in scalar value of anisotropy
\[ de^p, \Delta \] incremental deviatoric plastic strain
\[ d\gamma \] change in value of bonding
\[ [D]^\alpha \] elastic constitutive matrix
\[ \varepsilon^e, \varepsilon^\theta \] elastic and plastic strain vector
\[ f_y, \widehat{f}_y \] yield function and trial value of yield function
\[ G \] shear modulus
\[ H_1, H_2, H_3 \] hardening moduli
\[ \eta \] stress ratio
\[ K_0 \] lateral earth pressure at rest
\[ \kappa \] swelling index
\[ \lambda \] compression index
\[ \lambda_1 \] initial compression index
\[ \Lambda \] plastic multiplier
\[ M \] stress ratio at critical state
\[ \nu \] specific volume
\[ \varepsilon \] rate of destructuration
\[ \varepsilon_{sd} \] rate of destructuration due to deviator stress
\[ OCR \] over-consolidation ratio
\[ p' \] mean effective stress
\[ p_m' \] effective preconsolidation pressure
\[ \rho_m' \] size of intrinsic yield curve
\[ q \] deviatoric stress
\[ r_r, r_f \] residual errors
\[ RTOL, FTOL \] tolerance values
\[ \sigma' \] effective stress state
\[ \sigma_{ij} \] deviatoric effective stress vector
\[ \nu' \] poisson’s ratio
\[ \chi_0, \chi \] initial and current value of bonding
\[ \omega \] rate of rotation
\[ \omega_\beta \] rate of rotation due to deviator stress

(iteative) method. In the explicit method the solution is approached using the previously known stress point’s response in order to approximate the nonlinear material behaviour. This method is popular within the research community mainly due to its simplicity. However, the main issue in an explicit algorithm is that its success largely relies on the size of the incremental load steps, therefore the results particularly at boundary value level, can be load step dependent. This problem usually results in limiting the application of an advanced constitutive model at practical level. Nevertheless, there are modifications such as adopting an automatic substepping technique [14] to improve the performance of an explicit implementation. In the implicit or so-called return mapping algorithm [15] the constitutive model’s governing equations are treated nonlinearly and are solved in an iterative manner until the drift from the yield surface is small enough [16]. In this method the computed stress state automatically satisfies the yield condition. Also it does not need the computation of the intersection with the yield surface as the stress state changes from elastic to elasto-plastic [17]. The implicit method is particularly attractive if the model is to be used for practical simulations, which is mainly due to the fact that it is much more stable and efficient. However its application for advanced constitutive models is very difficult as it requires second order derivatives of the yield function.

In this study the return mapping algorithm [15] has been utilised for integrating the S-CLAY1S model. Previously, the applications of the return mapping algorithm to the modified Cam-Clay (MCC) model [18], through implicit and semi-implicit integration schemes, were investigated in [16,17]. It was found that the algorithm is capable of efficiently solving the nonlinear system of equations of the MCC model even during large strain increments. More recently Pipatpongsa et al. have shown that the form of the MCC’s yield function influences the performance of their developed backward Euler stress update algorithm for the model [19]; and, in case of original Cam-Clay model [20] the success or failure of their semi-implicit implementation has also been dependant on the yield function’s form [21]. In a separate work, Coombs and Crouch [22] have demonstrated the effects that alternative expressions of yield function has on the robustness and efficiency of a backward Euler stress integration method for a critical state based hyperplastic model.

In this paper a new Modified Newton–Raphson stress update scheme is employed for the numerical implementation of the S-CLAY1S model. The scheme has proven to have very high efficiency, robustness and accuracy for integrating the constitutive relationship during incremental elasto-plastic response [23]. It is also shown that the yield function form of the S-CLAY1S model can be presented with different alternative formulations; therefore, the corresponding solution algorithm results in different convergence speeds depending on which form of the yield function is used. Through model performance comparisons the most suitable formulation of the advanced anisotropic model is identified for its implementation.

2. S-CLAY1S model

The S-CLAY1S is an advanced critical state model which accounts for bonding and destructuration, in addition to plastic anisotropy [10]. In three-dimensional stress space the yield surface of the S-CLAY1S model, \( f_y \), is a sheared ellipsoid (see Fig. 1(a)) defined as

\[
 f_y = \frac{3}{2} \left[ (\sigma_d - p'\alpha_d)^2 (\sigma_d - p'\alpha_d) \right] - \left[ M^2 - \frac{3}{2}(\alpha_d)^2 (\alpha_d) \right] (p_m' - p') p' = 0
\]

(1)

In the above equation \( \sigma_d \) and \( \alpha_d \) are the deviatoric stress tensor and the deviatoric fabric tensor respectively, \( M \) is the critical state value, \( p' \) is the mean effective stress, and \( p_m' \) is the size of the yield surface related to the soil’s preconsolidation pressure. The effect of bonding in the S-CLAY1S model is described by an intrinsic yield surface [24] which has the same shape and inclination of the natural yield surface but with a smaller size. The size of the intrinsic yield surface is specified by parameter \( p_m' \) that is related to the size \( p_m' \) of the natural yield surface by parameter \( \chi \) as the current amount of bonding

\[
p_m' = (1 + \chi) p_m'
\]

(2)

S-CLAY1S model incorporates three hardening laws. The first of these, relates the increase in the size of the intrinsic yield surface to the increments of plastic volumetric strain (\( de^\nu \))

\[
dp_m' = \frac{\nu p_m'}{\lambda - \kappa} de^\nu
\]

(3)

where \( \nu \) is the specific volume, \( \lambda \) is the gradient of the intrinsic normal compression line in the compression plane (\( \ln p' - \nu \)).
The S-CLAY1S model, from which three different forms are summarised in [8] which describes the rotation of the yield surface with plastic straining. The second hardening law is the rotational hardening law, which describes the degradation of bonding with plastic straining. The third hardening law in S-CLAY1S is destructuration, which assumes that the total strain increment consists of elastic strain, plastic deviatoric strain, plastic volumetric strains in rotating the yield surface. The third hardening law in S-CLAY1S is destructuration, which describes the rotation of the yield surface toward its current target value, and the relative effectiveness of plastic deviatoric strains and plastic volumetric strains in rotating the yield surface.

\[ \text{Function form} \]

Table 1

<table>
<thead>
<tr>
<th>Alternative forms for the S-CLAY1S yield function, ( f_y ).</th>
</tr>
</thead>
<tbody>
<tr>
<td>Function form</td>
</tr>
<tr>
<td>( f_y^1 = \frac{2}{3} \left[ \sigma_d - p \alpha_d \right]^2 - \left( M^2 - \sigma_d^2 \right) p_{\text{an}} = p \prime - p = 0 )</td>
</tr>
<tr>
<td>( f_y^2 = \frac{2}{3} \left[ \sigma_d - p \alpha_d \right]^2 - \left( M^2 - \sigma_d^2 \right) p_{\text{an}} = 0 )</td>
</tr>
<tr>
<td>( f_y^3 = \frac{2}{3} \sigma_d - (M^2 - \sigma_d^2) p_{\text{an}}/C_0 \left( \sigma_d^2 - \sigma_{\prime}^2 \right) - c_d^2 = 0 )</td>
</tr>
</tbody>
</table>

where \( \sigma_d \) is the deviatoric stress tensor; \( p \) is the mean effective stress; \( \alpha_d \) is the deviatoric fabric tensor; \( \alpha \) is a scalar parameter for the special case of cross-anisotropy and its value is obtained as \( \alpha = (3/2)\left( \sigma_d^2(\alpha_d^2) \right) \); \( M \) is the value of the stress ratio at critical state; and \( p_{\text{an}} \) is the size of the yield surface.

The elastic behaviour in the model is formulated with the same flow rule for evaluating plastic strain increments, \( \Delta \varepsilon \), we have

\[ \Delta \varepsilon = \Delta \varepsilon^e + \Delta \varepsilon^p \] (6)

where \( \Delta \) remarks an incremental operator, and the underlined symbolises its vector form. Assuming an associated flow rule for evaluating plastic strain increments, \( \Delta \varepsilon^p \), we have

\[ \Delta \varepsilon = \Delta \varepsilon^e + \Delta \lambda \frac{\partial f_y}{\partial \sigma} \] (7 a)

hence the stress increment for a given trial strain can be obtained from

\[ \Delta \sigma^\prime = [\Delta \varepsilon^e + \Delta \lambda \frac{\partial f_y}{\partial \sigma}] \Delta \sigma \] (7 b)

where \( \Delta \lambda \) is the so-called plastic multiplier, \( \sigma^\prime \) is the effective stress and \([\Delta \varepsilon^e] \) is the plastic strain increment. To derive the plastic multiplier of the S-CLAY1S model, the yield function can be expanded with the Taylor series approach [23], by ignoring second or higher order terms we have

\[ f_y^\prime = f_y^e + \frac{\partial f_y}{\partial \sigma} \Delta \sigma^\prime + \frac{\partial f_y}{\partial p} \Delta p_{\text{an}} + \frac{\partial f_y}{\partial \alpha} \Delta \alpha + \frac{\partial f_y}{\partial \chi} \Delta \chi \] (7 c)

when \( f_y^\prime \) is the so-called trial yield function value. Given that the strain increment has all been applied to arrive at the elastic stress predictor, the MNR algorithm involves zero strain increment. Therefore Eq. (7-b) becomes

\[ \Delta \sigma^\prime = -[\Delta \varepsilon^e] \Delta \sigma^\prime \] (7 d)

The yield function value in Eq. (7-c) is zero as a stress state cannot stay outside the yield surface, therefore by using Eqs. (7-c)
and (7-d), the plastic multiplier is derived as

$$\Delta \Lambda = -f_y \frac{\partial \sigma_n}{\partial \sigma_n} \{ -f_y (\frac{\partial f_y}{\partial \sigma_n}) + H_1 + H_2 + H_3 \}$$  \hspace{1cm} (8 - a)

where

$$H_1 = \frac{\partial f_y}{\partial \sigma_m} \frac{\partial \sigma_m}{\partial \sigma_n} \frac{\partial f_y}{\partial \sigma_n}$$  \hspace{1cm} (8 - b)

$$H_2 = \left( \frac{\partial \sigma_n}{\partial \sigma_n} \right)^T \left[ \frac{\partial \sigma_n}{\partial \sigma_n} \frac{\partial f_y}{\partial \sigma_n} + \frac{\partial \sigma_n}{\partial \sigma_n} \frac{\partial f_y}{\partial \sigma_n} \right] \left( \frac{\partial \sigma_n}{\partial \sigma_n} \right)^T$$  \hspace{1cm} (8 - c)

$$H_3 = \frac{\partial f_y}{\partial \sigma_n} \left[ \frac{\partial \sigma_n}{\partial \sigma_n} \frac{\partial f_y}{\partial \sigma_n} + \frac{\partial \sigma_n}{\partial \sigma_n} \frac{\partial f_y}{\partial \sigma_n} \right]$$  \hspace{1cm} (8 - d)

Dimensional analysis of Eq. (7-a) is required to determine whether it is correct to use any of the three yield function forms, presented in Table 1, in the MNR algorithm. The first two terms of the Eq. (7-a) are strains and therefore dimensionless, hence the third term, i.e. the plastic strain increment, must also be dimensionless. The results of the dimensional analysis are summarised in Table 2 with SI units used in the analyses.

As it is seen in the table, all three yield functions result in dimensionless plastic strain increment, therefore they are all dimensionally consistent and suitable to be used in the MNR process. The two sets of residual equations in the developed MNR algorithm are

$$r_e = \Delta \varepsilon - \Delta \varepsilon^{e^{r}} - \Delta \Lambda \frac{\partial f_y}{\partial \sigma_n}$$  \hspace{1cm} (9 - a)

and

$$r_f = f_y(\sigma_n^{e^{r}})$$  \hspace{1cm} (9 - b)

Differentiating the residuals in equations above with regards to the stresses leads a set of linearised equations which, after neglecting higher order terms, can be written in an iterative manner as follows

$$\Delta \varepsilon_n - \Delta \varepsilon_n^{e^{r}} - \delta \Delta \varepsilon_n^{e^{r}}_n - \Delta \Lambda_0 \frac{\partial f_y}{\partial \sigma_n} - \Delta \Lambda_i \frac{\partial ^2 f_y}{\partial \sigma_n^2} \delta \sigma_n^{e^{r}} - \Delta \Lambda_{n+1} \frac{\partial f_y}{\partial \sigma_n} = 0$$  \hspace{1cm} (10 - a)

$$f_y + \frac{\partial f_y}{\partial \sigma_n} \delta \sigma_n^{e^{r}} = 0$$  \hspace{1cm} (10 - b)

Eq. (10) can be expressed in the matrix form as

$$\begin{bmatrix} I + \Delta \Lambda_0 \frac{\partial f_y}{\partial \sigma_n} \frac{\partial \sigma_n^{e^{r}}}{\partial \sigma_n} \biggr|_{n+1} & \frac{\partial f_y}{\partial \sigma_n} \biggr|_{n+1} \delta \Lambda_{n+1} \biggr] \delta \Delta \varepsilon_n^{e^{r}} = r_e \end{bmatrix}$$

$$\begin{bmatrix} I + \Delta \Lambda_0 \frac{\partial f_y}{\partial \sigma_n} \frac{\partial \sigma_n^{e^{r}}}{\partial \sigma_n} \biggr|_{n+1} & \frac{\partial f_y}{\partial \sigma_n} \biggr|_{n+1} \delta \Lambda_{n+1} \biggr] \delta \Delta \varepsilon_n^{e^{r}} = r_e \end{bmatrix}$$ \hspace{1cm} (11)

where $I$ is the fourth order identity tensor. The above system of equations is solved in an iterative manner, and as the result the changes of elastic strain increment and the plastic multiplier will be obtained. The state variables of the model are updated in every iteration, and the procedure of updating $\Delta \varepsilon$ and $\Delta \Lambda$ within a load step is repeated until the norms of both residuals, i.e. $r_e$ and $r_f$ are less than the prescribed tolerances, namely $RTOL$ and $FTOL$. Now the trial elastic strain increment, $\Delta \varepsilon e^{r}$, is defined by the following equation where $\Delta \varepsilon e^{r}$ is the initial elastic strain increment converged in the previous step and $\delta \Delta \varepsilon e^{r}$ is the changes of elastic strain increment obtained from the above system of equations

$$\Delta \varepsilon e^{r} = \Delta \varepsilon e^{r} + \delta \Delta \varepsilon e^{r}$$  \hspace{1cm} (12)

Accordingly the trial value of plastic multiplier gets updated too. The yield function, $f_y$, appears in Eq. (11) can be replaced by individual yield functions listed in Table 1.

4. Computational performance

Different formulations of the S-CLAY1S yield function are all representing similar equation when equaling to zero, therefore they are supposed to produce unique surfaces. However, in the triaxial stress space $f_y$ results in numerical singularity when mean effective stress (i.e. $\sigma'$) closes to zero (see Fig. 2) which simply is due to the presence of $\sigma'$ in the denominator term of the equation (see Table 1).

It is commonly known that critical state models do not have any strength if the effective mean stress is zero or tensile, therefore for them to be successfully used a compressive initial stress state is required to be established at the point of interest in the numerical domain. Taking this into account, there is not a particular difficulty in considering $f_y$ in the model implementation and its application. In the following, assessments of the proposed MNR numerical algorithm for S-CLAY1S integration with regards to each alternative form of the yield function will be presented. Because the S-CLAY1S model is already validated (e.g., [11–13]), in this paper only the numerical performance of the proposed algorithm is studied, and the effect of the yield function equation on the proposed stress update scheme is investigated.

4.1. At stress point level

A set of parameter values for the S-CLAY1S constitutive model representing a well-known soft clay, namely Bothkennar clay, is selected in order to demonstrate the computational performance of the model implementation (see Table 3). Bothkennar clay is a soft normally consolidated marine clay deposited on the Forth River Estuary which is located approximately midway between Glasgow and Edinburgh in Scotland. There is extensive laboratory data available for Bothkennar clay (e.g., [26]) which makes it possible to derive a consistent set of material parameters for the advanced constitutive model being studied. The details of parameter determination can be found in [26,27].

At first stage, the verification of the implementation is performed by using an in-house developed integration point programme (IPP). The S-CLAY1S model subroutine is strain driven, i.e. known increments in strain are input to the model and the
corresponding stresses are output. Thus, by specifying strain increments and the material parameters, the corresponding stress output can be used for evaluating model performance without the need for any finite element (FE) software. Depending on which form of the yield function is employed in the implementation, the convergence speed of the iterative scheme would vary as it depends on the degree of nonlinearity in the system of governing equations.

Different combinations of strain loading have been considered in the way that they cover many important stress–strain conditions in common soil element tests. In each case an identical initial stress inside the yield surface was selected and then under a specific loading condition the stress path was monitored. The magnitudes of the imposed strains were selected in the way that they cover a significant range of loading conditions. A summary of the applied strains is shown in the first column of Table 4. The performance of the implementation using different functions can be studied by analysing their efficiency which can be represented by the number of iterations required for convergence. As it is seen in Fig. 2, for the majority of loading cases, the employment of \( f_y^1 \) results in the lowest number of iterations. Yield function equations \( f_y^1 \) and \( f_y^2 \) generally seem to produce very similar number of iterations. Table 4 also shows the required CPU time (in milliseconds) for each analysis. The measured computational times for the analyses are based on a PC with an Intel Core i7-3820 3.60 GHz processor, and they are recorded using the standard SYSTEM_CLOCK subroutine [28] in Lahey/Fujitsu FORTRAN Professional V7.3 compiler. Even though in real-time scale the durations of analyses are almost indifferent; however, in computational sense \( f_y^2 \) is evidently cheaper as compared to the other two forms of the yield function, its application results in solutions being converged within less CPU time.

Stress return paths for cases representing more commonly observed conditions in soil tests are presented in Fig. 3. Among them, Fig. 3(a)–(c) represents triaxial undrained compression, undrained extension and isotropic compression conditions respectively, and Fig. 3(d) represents oedometer condition (one dimensional straining).

It is seen that in all cases of different given strain increments, the return mappings of the arbitrary trial stresses converge to identical final stresses on the yield surface; that is an indication of the robustness of developed MNR algorithm. However, the return mapping paths and the required number of iterations vary depending on the form of the yield function. This is mainly because the direction of stress return path depends on different derivations of the yield function (with regards to stresses and/or state variables, assuming associated flow rule) which are dissimilar when different yield function equations are employed. From the results, it is noteworthy to mention that in the first iteration \( f_y^1 \) employment usually produces the most erroneous stress value compared to when the other two forms are used (see Figs. 3(a), (c) and (d)).

In order to obtain more insight into the possible sensitivity of the MNR algorithm to the variations of the yield surface shape, a simple sensitivity analysis is likely to be helpful. Based on the authors’ experience the S-CLAY1S model performance is not particularly sensitive to different, but yet appropriate, combinations of constant values in rotational and destructuration hardening laws, namely \( \alpha, \beta, \zeta \) and \( \zeta_d \). Therefore, for the sake of simplicity, the sensitivity analysis has only been performed within an upper and lower plausible limit for state variables \( \alpha_0 \) and \( X_0 \). Four different series of analyses have been done with changing each of \( \alpha_0 \) and \( X_0 \) values by \( \pm 25\% \) around their representative values listed in Table 3. Only the four stress paths illustrated in Fig. 3 are simulated and the results of the analyses are summarised in Tables 5–8. Looking at the results in these tables and comparing

### Table 3

<table>
<thead>
<tr>
<th>( \varepsilon_y )</th>
<th>OCR</th>
<th>( \alpha_0 )</th>
<th>( X_0 )</th>
<th>( \kappa )</th>
<th>( \nu )</th>
<th>( \lambda )</th>
<th>( M )</th>
<th>( \omega )</th>
<th>( \alpha_0 )</th>
<th>( \zeta )</th>
<th>( \zeta_d )</th>
</tr>
</thead>
<tbody>
<tr>
<td>2.0</td>
<td>1.5</td>
<td>0.59</td>
<td>8.0</td>
<td>0.02</td>
<td>0.2</td>
<td>0.18</td>
<td>1.5</td>
<td>50</td>
<td>1.0</td>
<td>9</td>
<td>0.2</td>
</tr>
</tbody>
</table>

### Table 4

<table>
<thead>
<tr>
<th>Strain increment</th>
<th>( f_y^1 )</th>
<th>Iter.</th>
<th>( t ) (ms)</th>
<th>( f_y^2 )</th>
<th>Iter.</th>
<th>( t ) (ms)</th>
<th>( f_y^3 )</th>
<th>Iter.</th>
<th>( t ) (ms)</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \varepsilon_y = -0.50% ) and ( \varepsilon_y = -0.25% )</td>
<td>10</td>
<td>4</td>
<td>10</td>
<td>3</td>
<td>8</td>
<td>2</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \varepsilon_y = 0.50% ) and ( \varepsilon_y = -0.25% )</td>
<td>10</td>
<td>2</td>
<td>10</td>
<td>2</td>
<td>9</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \varepsilon_y = 0.50% ) and ( \varepsilon_x = 0.0 )</td>
<td>9</td>
<td>2</td>
<td>9</td>
<td>2</td>
<td>8</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \varepsilon_y = 0.50% ) and ( \varepsilon_x = 0.25% )</td>
<td>12</td>
<td>3</td>
<td>11</td>
<td>2</td>
<td>11</td>
<td>2</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \varepsilon_y = 0.50% ) and ( \varepsilon_x = 0.05% )</td>
<td>10</td>
<td>2</td>
<td>10</td>
<td>2</td>
<td>8</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>( \varepsilon_y = 0.50% ) and ( \varepsilon_x = 0.10% )</td>
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<td>10</td>
<td>2</td>
<td>8</td>
<td>1</td>
<td></td>
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<tr>
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<td>3</td>
<td>10</td>
<td>2</td>
<td>10</td>
<td>1</td>
<td></td>
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<tr>
<td>( \varepsilon_y = 0.50% ) and ( \varepsilon_x = 0.30% )</td>
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<td>3</td>
<td>11</td>
<td>2</td>
<td>12</td>
<td>2</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \varepsilon_y = 0.50% ) and ( \varepsilon_x = 0.40% )</td>
<td>14</td>
<td>3</td>
<td>14</td>
<td>2</td>
<td>13</td>
<td>2</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \varepsilon_y = 0.50% ) and ( \varepsilon_x = 0.50% )</td>
<td>15</td>
<td>4</td>
<td>15</td>
<td>3</td>
<td>14</td>
<td>2</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Fig. 2. Contours of arbitrary S-CLAY1S yield surfaces in triaxial stress space using (a) \( f_y^1 \) or \( f_y^3 \), and (b) \( f_y^2 \).
Table 5
Sensitivity of the Newton–Raphson stress update scheme to the variations of the yield surface shape by a 25% increase in the $\alpha_0$ value.

<table>
<thead>
<tr>
<th>Strain increment</th>
<th>$f_y^1$</th>
<th>$f_y^2$</th>
<th>$f_y^3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = -0.25%$</td>
<td>10 / 4</td>
<td>10 / 3</td>
<td>8 / 2</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = -0.25%$</td>
<td>9 / 2</td>
<td>9 / 3</td>
<td>9 / 1</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = 0.0%$</td>
<td>10 / 2</td>
<td>9 / 2</td>
<td>8 / 1</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = 0.50%$</td>
<td>15 / 4</td>
<td>16 / 3</td>
<td>15 / 2</td>
</tr>
</tbody>
</table>

Table 6
Sensitivity of the Newton–Raphson stress update scheme to the variations of the yield surface shape by a 25% reduction in the $\alpha_0$ value.

<table>
<thead>
<tr>
<th>Strain increment</th>
<th>$f_y^1$</th>
<th>$f_y^2$</th>
<th>$f_y^3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\varepsilon_x = -0.50%$ and $\varepsilon_y = -0.25%$</td>
<td>11 / 4</td>
<td>10 / 3</td>
<td>8 / 2</td>
</tr>
<tr>
<td>$\varepsilon_x = -0.50%$ and $\varepsilon_y = -0.25%$</td>
<td>9 / 2</td>
<td>9 / 3</td>
<td>9 / 1</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.0%$ and $\varepsilon_y = 0.0%$</td>
<td>9 / 2</td>
<td>9 / 3</td>
<td>8 / 1</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = 0.50%$</td>
<td>14 / 3</td>
<td>14 / 3</td>
<td>14 / 2</td>
</tr>
</tbody>
</table>

Table 7
Sensitivity of the Newton–Raphson stress update scheme to the variations of the yield surface shape by a 25% increase in the $\chi_0$ value.

<table>
<thead>
<tr>
<th>Strain increment</th>
<th>$f_y^1$</th>
<th>$f_y^2$</th>
<th>$f_y^3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\varepsilon_x = -0.50%$ and $\varepsilon_y = -0.25%$</td>
<td>10 / 4</td>
<td>10 / 3</td>
<td>8 / 2</td>
</tr>
<tr>
<td>$\varepsilon_x = -0.50%$ and $\varepsilon_y = -0.25%$</td>
<td>10 / 2</td>
<td>9 / 2</td>
<td>9 / 1</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.0%$ and $\varepsilon_y = 0.0%$</td>
<td>9 / 2</td>
<td>9 / 2</td>
<td>8 / 1</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = 0.50%$</td>
<td>15 / 4</td>
<td>15 / 3</td>
<td>14 / 2</td>
</tr>
</tbody>
</table>

Table 8
Sensitivity of the Newton–Raphson stress update scheme to the variations of the yield surface shape by a 25% reduction in the $\chi_0$ value.

<table>
<thead>
<tr>
<th>Strain increment</th>
<th>$f_y^1$</th>
<th>$f_y^2$</th>
<th>$f_y^3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = 0.50%$</td>
<td>11 / 4</td>
<td>10 / 3</td>
<td>8 / 2</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = 0.50%$</td>
<td>10 / 2</td>
<td>9 / 2</td>
<td>9 / 1</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = 0.50%$</td>
<td>9 / 2</td>
<td>9 / 3</td>
<td>8 / 1</td>
</tr>
<tr>
<td>$\varepsilon_x = 0.50%$ and $\varepsilon_y = 0.50%$</td>
<td>15 / 4</td>
<td>15 / 3</td>
<td>14 / 2</td>
</tr>
</tbody>
</table>

Fig. 3. Return-mapping paths for four different sets of imposed strain increments.

To compare the numerical performance of the implicit integration scheme described in the previous section at boundary value level, a relatively simple example is analysed with finite element method. The boundary value problem involves plane strain analysis of a hypothetical embankment constructed on a soft soil deposit with the properties of Bothkennar clay. This is a good case for assessing the performance of the integration algorithm with different yield function forms due to the significant rotation of principal stresses particular in areas directly below the embankment. The S-CLAY15 model has been implemented into PLAXIS finite element programme as a user-defined soil model in
In order to be used for the boundary value level study. It is assumed that the embankment is 2 m high, its width at the top is 10 m and the side slopes have a gradient of 1:2. A simplified soil profile is considered for the soft deposit underneath the embankment, which includes a 1 m deep over-consolidated dry crust followed by 29 m of soft Bothkennar clay. The groundwater table is located 1 m below the ground surface. A schematic view of the geometry of the embankment and the soft soil profile underneath it are shown in Fig. 4.

In terms of boundary conditions, the left and right boundaries of the model are fixed in horizontal direction and the bottom boundary is fixed in both horizontal and vertical directions. Taking advantage of the symmetry of the problem, just half of the geometry is considered for the FE analysis. The simulations have been done using 1696, 15-noded triangular elements with 12 stress points, and mesh-sensitivity studies have been performed to ensure the accuracy of the results.

The embankment is constructed in two steps; at each step 1 m of fill material is placed during 5 days. The embankment which is assumed to be made of granular fill is modelled with a simple Mohr–Coulomb model considering the following typical values for the embankment material: $E=40,000$ kN/m$^2$, $\nu=0.35$, $q_f=40$, $\psi_f=0$, $c_f=2$ kN/m$^2$, and $\gamma=20$ kN/m$^3$. Mohr–Coulomb model is also used to represent the behaviour of the over-consolidated dry crust layer, assuming the following relevant parameter values: $E=3000$ kN/m$^2$, $\nu=0.2$, $q_f=37.1$, $\psi_f=0$, $c_f=2$ kN/m$^2$, and $\gamma=19$ kN/m$^3$. The S-CLAY1S material parameters used for the Bothkennar clay are the same as those used for the stress point study. The permeability, $k$, of the clay is assumed to be isotropic and equal to 2.5E-4 (m/day). The FE analysis includes two construction phases followed by a consolidation phase that continues until full dissipation of excess pore pressures in the soft soil layer. It should also be added that for the example considered here, the initial state of stress is generated by adopting $K_0$-procedure [30] where $K_0$, which is the in-situ coefficient of earth pressure at rest, is 0.5. The existence of dry crust layer, modelled with a simple Mohr–Coulomb model, ensures that all stress points with the critical state model S-CLAY1S start from a positive stress state.

At first instance, the finite element analysis of the embankment is repeated three times, each time using one form of S-CLAY1S yield function for the user defined model implementation. For simplicity, here they are referred to as Case I, Case II and Case III where $f_1$ is used in Case I, and $f_2$ and $f_3$ are used in Case II and Case III respectively. In all FE runs the stresses during plastic straining are restored to the yield surface using an absolute residual tolerance value of $1 \times 10^{-8}$ for both RTOL and FTOL. At the end of all analysis 4037 stress points inside the soft soil domain become plastic (see Fig. 5).

Table 9 shows the results of the three analyses in terms of total CPU time, total number of required MNR iterations, maximum number of required MNR iterations in one load step, average number of required MNR iterations for all load steps, average value of all residuals for $r_f$, and average value of all residuals for $r_r$. Similar to the simulations at stress point level, all the timing results presented in this section are also for a PC with Intel Core i7-3820 3.60 GHz processor with the Lahey/Fujitsu FORTRAN Professional V7.3 compiler. The CPU time recorded is for the entire finite element analysis, not just the stress integration part, as this is of more relevance to the design utilisation of numerical models.

Comparing the results in Table 9, it can be seen that the duration of FE analysis for Case III is noticeably less than the time required for the other two cases; it is almost 21% shorter than Case I, which is remarkable. Case II also results in almost 11% saving in computational cost, compared to Case I. Therefore, it is not surprising to see that overall Case I requires more MNR iterations for stress integration during plastic straining. In addition, while the values

![Fig. 4. Geometry of the boundary value problem and the idealised soil profile.](image)

![Fig. 5. Plastic zone within the soft soil layer, the location of the monitored stress point, and the location of point A for deformation measurement.](image)

<table>
<thead>
<tr>
<th>Yield function</th>
<th>CPU time (s)</th>
<th>Total req. MNR iter.</th>
<th>Max. req. MNR iter.</th>
<th>Avg. req. MNR iter.</th>
<th>Avg. $r_f$ ($f_f$ residual)</th>
<th>Avg. $r_r$ ($\varepsilon_f$ residual)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Case I: using $f_1$</td>
<td>681</td>
<td>20,919,493</td>
<td>14</td>
<td>9.19</td>
<td>$2.72 \times 10^{-9}$</td>
<td>$2.92 \times 10^{-14}$</td>
</tr>
<tr>
<td>Case II: using $f_2$</td>
<td>607</td>
<td>18,535,346</td>
<td>13</td>
<td>8.27</td>
<td>$2.79 \times 10^{-9}$</td>
<td>$4.25 \times 10^{-13}$</td>
</tr>
<tr>
<td>Case III: using $f_3$</td>
<td>538</td>
<td>16,319,311</td>
<td>9</td>
<td>6.39</td>
<td>$2.65 \times 10^{-9}$</td>
<td>$1.19 \times 10^{-12}$</td>
</tr>
</tbody>
</table>
of both \( RTOL \) and \( FTOL \) tolerances are the same, in general after convergence the average \( r_e \) residuals are several orders of magnitude smaller than average \( r_f \) residuals; this indicates that in the concurrent iterative procedure, the tolerance with regards to the strain residual is always satisfied first, but satisfying \( FTOL \) requires further iterations in order to achieve the required error control, which subsequently results in \( r_e \) getting substantially smaller than the desired \( RTOL \).

For an enhanced evaluation of the MNR performance with different yield function forms, the error proportionality within the plastic domain of Fig. 5 are presented in Figs. 6–8 using contour maps of normalised residuals and iteration numbers. It should be mentioned that these contours are plotted at the end of consolidation in each analysis and hence signify values only at the final time increment of the analyses. However, they still provide a very good graphical glimpse of the return mapping algorithm function in different cases.

From Figs. 6–8 it is generally observed that residual contour maps are somehow representing the mobilised slip surface due to loading (particularly in Cases II and III); in Fig. 8 related to Case III even the stress concentration below the embankment toe can be identified. This could be expected given the fact that stress points along the slip surface are the more critical ones where the imposed stress distribution deviates significantly from the initial stress distribution in the ground. Case I results in less residual values, i.e. less errors, over the plastic points; however that is accompanied with higher computational cost as in this particular snapshot for the majority of the points 13 stress integration

![Contour maps for normalised residuals \( r_f \) and \( r_e \), and required number of iterations based on \( f_1^y \), Case I; \( (r_f)_{\text{max}} \) is 9.92 \( \times 10^{-10} \) and \( (r_e)_{\text{max}} \) is 5.59 \( \times 10^{-14} \).](image)

![Contour maps for normalised residuals \( r_f \) and \( r_e \), and required number of iterations based on \( f_2^y \), Case II; \( (r_f)_{\text{max}} \) is 9.94 \( \times 10^{-10} \) and \( (r_e)_{\text{max}} \) is 6.65 \( \times 10^{-13} \).](image)
iterations are needed (see Fig. 6). Towards Case II and Case III the errors become more widely distributed between maximum and minimum amounts (see Figs. 7 and 8), and the required MNR iterations in the plastic domain reduces considerably in Case III.

For further comparison, the above finite element analysis has been repeated with tolerance values, RTOL and FTOL, relaxed once to $1 \times 10^{-6}$ and the other time to $1 \times 10^{-4}$. For all nine analyses, the MNR iteration numbers and the values of residuals were monitored at a single stress point inside element number 359 within the plastic domain (see Fig. 5). The results of all FE analyses are summarised in Tables 10–12.

The results from all runs show remarkably consistent trend; indeed, the same observations and conclusions, as those explained above based on the results in Table 9, apply. All analyses display a “tolerance proportionality” [31] with regards to FTOL; also, the required number of MNR iterations and hence the analysis duration is constantly decreasing when from $f_y^1$ is used to when $f_y^2$ is used. In addition, it is seen that in each case the timings are largely sensitive to the specified error tolerances. For example, the difference in the CPU time related to the employment of $f_y^2$, between when $FTOL=RTOL=10^{-8}$ and when $FTOL=RTOL=10^{-4}$, is more than 34% (see Table 10). Also significant variations in the number of required MNR iterations is observed, for example when yield function form $f_y^1$ is used, the average number of required MNR iterations with $FTOL=RTOL=10^{-4}$ is over 45 % less than the same iteration number when $FTOL=RTOL=10^{-8}$ (see Table 11). It should be noted that in all nine analyses, settlement has been monitored at the ground surface under the centreline of the embankment (point A in Fig. 5), and in all runs, but one, the predicted displacements have been uniformly matching. The one analysis with dissimilar settlement prediction is related to the employment of $f_y^2$ with $FTOL=RTOL=10^{-4}$ where the final settlement has been 0.8 mm less than other FE runs; this appears to be a negligible variance.

5. Conclusion

The robustness of a modified Newton–Raphson algorithm applied to the S-CLAY1S model has been evaluated using different forms of its yield function. Although the proposed stress update scheme is generally very robust, and impartial in terms of accuracy; however, its convergence performance is found to be

---

Table 10

<table>
<thead>
<tr>
<th>Yield function: $f_y^1$</th>
<th>CPU time (s)</th>
<th>Total req. MNR iter.</th>
<th>Max. req. MNR iter.</th>
<th>Avg. req. MNR iter.</th>
<th>Avg. $r_f$ (strain residual)</th>
<th>Avg. $r_r$ (residual)</th>
</tr>
</thead>
<tbody>
<tr>
<td>$FTOL=RTOL=10^{-8}$</td>
<td>681</td>
<td>8444</td>
<td>9</td>
<td>8.32</td>
<td>$9.93 \times 10^{-9}$</td>
<td>$1.98 \times 10^{-14}$</td>
</tr>
<tr>
<td>$FTOL=RTOL=10^{-6}$</td>
<td>553</td>
<td>6701</td>
<td>8</td>
<td>6.59</td>
<td>$3.72 \times 10^{-7}$</td>
<td>$2.66 \times 10^{-12}$</td>
</tr>
<tr>
<td>$FTOL=RTOL=10^{-4}$</td>
<td>538</td>
<td>5359</td>
<td>6</td>
<td>5.27</td>
<td>$2.45 \times 10^{-5}$</td>
<td>$1.88 \times 10^{-10}$</td>
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Table 11

<table>
<thead>
<tr>
<th>Yield function: $f_y^2$</th>
<th>CPU time (s)</th>
<th>Total req. MNR iter.</th>
<th>Max. req. MNR iter.</th>
<th>Avg. req. MNR iter.</th>
<th>Avg. $r_f$ (strain residual)</th>
<th>Avg. $r_r$ (residual)</th>
</tr>
</thead>
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<tr>
<td>$FTOL=RTOL=10^{-8}$</td>
<td>607</td>
<td>7520</td>
<td>8</td>
<td>7.42</td>
<td>$2.60 \times 10^{-9}$</td>
<td>$2.16 \times 10^{-15}$</td>
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<td>$FTOL=RTOL=10^{-6}$</td>
<td>488</td>
<td>5862</td>
<td>7</td>
<td>5.78</td>
<td>$4.99 \times 10^{-7}$</td>
<td>$3.59 \times 10^{-11}$</td>
</tr>
<tr>
<td>$FTOL=RTOL=10^{-4}$</td>
<td>399</td>
<td>4505</td>
<td>5</td>
<td>4.43</td>
<td>$2.14 \times 10^{-5}$</td>
<td>$3.11 \times 10^{-9}$</td>
</tr>
</tbody>
</table>

Table 12

<table>
<thead>
<tr>
<th>Yield function: $f_y^3$</th>
<th>CPU time (s)</th>
<th>Total req. MNR iter.</th>
<th>Max. req. MNR iter.</th>
<th>Avg. req. MNR iter.</th>
<th>Avg. $r_f$ (strain residual)</th>
<th>Avg. $r_r$ (residual)</th>
</tr>
</thead>
<tbody>
<tr>
<td>$FTOL=RTOL=10^{-8}$</td>
<td>538</td>
<td>6555</td>
<td>7</td>
<td>6.46</td>
<td>$2.55 \times 10^{-9}$</td>
<td>$9.77 \times 10^{-13}$</td>
</tr>
<tr>
<td>$FTOL=RTOL=10^{-6}$</td>
<td>440</td>
<td>4902</td>
<td>6</td>
<td>4.82</td>
<td>$4.06 \times 10^{-7}$</td>
<td>$1.46 \times 10^{-10}$</td>
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<tr>
<td>$FTOL=RTOL=10^{-4}$</td>
<td>356</td>
<td>3564</td>
<td>4</td>
<td>3.54</td>
<td>$2.05 \times 10^{-5}$</td>
<td>$1.27 \times 10^{-8}$</td>
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</tbody>
</table>

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Fig. 8. Contour maps for normalised residuals $r_f$ and $r_r$, and required number of iterations based on $f_y^2$, Case III; $(r_f)_{\text{max}}$ is $9.97 \times 10^{-9}$ and $(r_r)_{\text{max}}$ is $1.98 \times 10^{-12}$. 

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considerably sensitive to the yield function equation as well as acceptable tolerances. There is a particular form of the yield function, i.e. $f_\phi$, that results in the smallest number of iterations and the most robust yield form. Over a simple example it was shown that employment of this form of the S-CLAY1S yield function equation for boundary value level analysis can result in over 20% saving in computational cost which is considerable. This is specifically promising given that the plastic domain in the analysed example was relatively small. Similar form of the yield function for other models can be of particular benefit for analysing highly non-linear problems where small load steps produce large strain increments, hence fewer stress update iterations are favourable to provide an economical measure for integrating elastoplastic constitutive laws. Overall this paper highlights the possibility of using simple numerical techniques to enhance the stress integration algorithm's performance for complex constitutive models particularly those belonging to the critical state family.

Acknowledgements

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References